# Using 3D affordance for grasping pose generation

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#### Outline

Motivation

What is Affordance?

Background

Current models:

- Implicit Estimation and Visual Affordance
- Language-Conditioned Affordance-Pose Detection
- Language Guided Affordance (AffordDexGrasp)

Summary

#### Motivation

Where would you grasp this teapot?



#### What is Affordance

"Affordance" depends on:

Object shape

Object type (what is it)

Intention (what do you want to do?)

Also, your manipulator

> Object in its context!How an object can be used



## Challenges:



## Old and new approaches

LEGACY APPROACHES	STILL IN USE	NEW RESEARCH
2D images	Depth cameras	6-DoF grasp pose estimation
Bounding boxes	Point clouds	Language guided
Masks	Pixel/point-based ground	Neural networks
Semantic recognition/	truths	Some combination of
segmentation	Random grasp generation	affordance and pose generation("End-to-End")

Input: point cloud (depth camera render)

Parallel processing of affordance and grasp detection

Combination at the end

Defined "mini-task actions" (set of affordances)

• E.g. wrap, grasp, pour, cut etc.

Confidence scores and affordance map are used to determine gasp candidate

new dataset: simulated ShapeNet with task-oriented category's

Labeling: binary (successful or not)





#### ADVANTAGE

#### DISADVANTAGE

Only image (point cloud) as input

Can evaluate and determine different affordances

Closed set

No natural language guidance

Not easily expandable to new objects/affordances

## Language-Conditioned Affordance-Pose Detection - Dataset

New dataset based on 3D AffordanceNet and 6-DoF GraspNet

Manual annotation of affordance category's

Converted to point cloud

Dataset is triplet of point could, affordance category and pose

# Language-Conditioned Affordance-Pose Detection

Input: point cloud and text (affordance)

Point cloud and text are encoded separately

Grasp estimation is diffusion based

Context for diffusion is made up of embeddings of:

• Point cloud, text, robot state and timestep

## Language-Conditioned Affordance-Pose Detection



# Language-Conditioned Affordance-Pose Detection

ADVANTAGE

Open vocabulary

Expansion to new affordances and objects possible

DISADVANTAGE

Always requires affordance instruction Expansion to new affordances limited Word choice matters Single object only

## AffordDexGrasp: Dataset

#### New open set dataset based on language-guided dexterous grasp dataset tabletop environment and 33 categories





Use the teapot with grasping the handle from the right.



Use the mug by contacting the handle from the right.



Unseen Category

Hold the body of trigger sprayer from the left.



Use the hammer and grasp handle from the back.



Hold the body of trigger sprayer from the left.









Use the frying pan and grasp handle from the back.



Hold the body of the lotion pump from the left.





Hold the bottle body from the left.





Use the frying pan and grasp handle from the right.





Use the mug by contacting the handle from the right.



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## AffordDexGrasp:

Input:

- Natural language user input
- RGB image
- Point cloud

Utilizes MLLM to intuit and guide affordance

Closer to sequential processing:

- First create an affordance map
- Then determine gasp from affordance map + affordance encoding

#### AffordDexGrasp:



#### AffordDexGrasp:

ADVANTAGE

Open to natural language instructionsBut not necessary!

Use of RGB image

Robust to different formulations

DISADVANTAGE

Input needs RBG image, point cloud and instruction for best performance

Still has real world gap

## Summary

Utilize 3D/2.5D vision

Gather context information (image + text/situation)

Integration of affordance and grasp pose estimation

Challenges:

- Pressley defining the problem (+dataset)
- Measuring success
- Applying to real world (multiple objects, point cloud render etc.)

#### Sources

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